

5-21-2013

# The Application of Singular Value Decomposition to Determine the Sources of Far Field Diesel Engine Noise

J Stuart Bolton

*Purdue University*, bolton@purdue.edu

Patricia Davies

*Purdue University*, davies@purdue.edu

Michael D. Hayward

*Purdue University*

Follow this and additional works at: <http://docs.lib.purdue.edu/herrick>

---

Bolton, J Stuart; Davies, Patricia; and Hayward, Michael D., "The Application of Singular Value Decomposition to Determine the Sources of Far Field Diesel Engine Noise" (2013). *Publications of the Ray W. Herrick Laboratories*. Paper 102.  
<http://docs.lib.purdue.edu/herrick/102>

This document has been made available through Purdue e-Pubs, a service of the Purdue University Libraries. Please contact [epubs@purdue.edu](mailto:epubs@purdue.edu) for additional information.

# **THE APPLICATION OF SINGULAR VALUE DECOMPOSITION TO DETERMINE THE SOURCES OF FAR FIELD DIESEL ENGINE NOISE**

Michael D. Hayward

May 21 2013

Patricia Davies

SAE Noise and Vibration Conference

J. Stuart Bolton

Grand Rapids, Michigan

# Introduction

- Demand for quieter engines is a constant driving force behind creation of competitive engines
- Determination of dominant noise sources in both the near- and far-fields within an engine is an integral step in development of quieter engines
- A method to assist in noise source identification in the near- and far-fields was desired to reduce the number of time-consuming and expensive fired and motored tests required



# Outline

Multiple Input/Multiple  
Output (MIMO) System



Transfer Path Estimation  
Between Inputs and Outputs

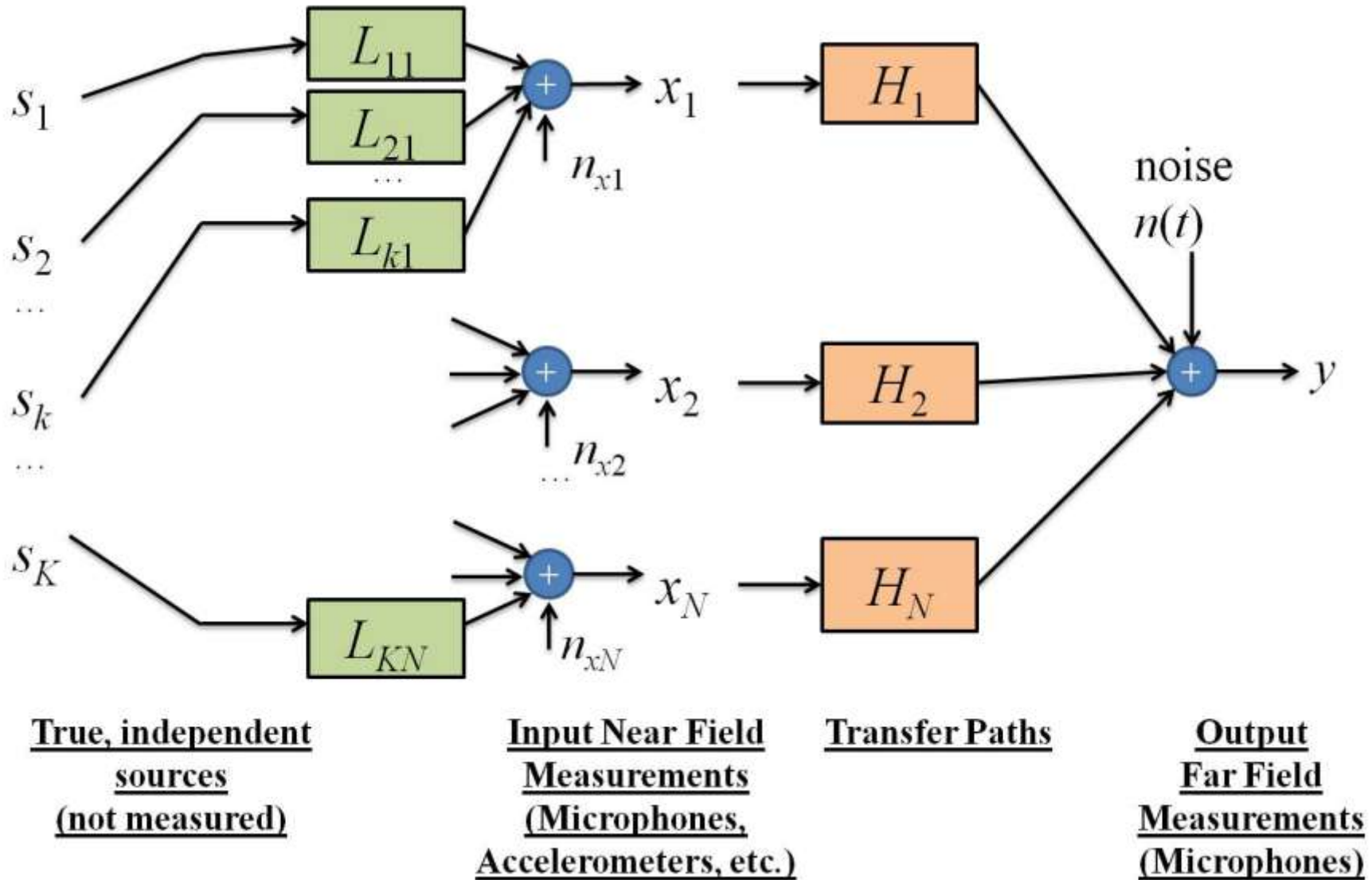


Singular Value Decomposition



SVD Contributions to Near-  
and Far-Field Measurements

# Multiple Input/Single Output System



# Solution of Cross-Spectral Matrix Problem

- A method to calculate transfer paths without including repeated information was required.
- In the following simplified equation,  $H_1$  and  $H_2$  are needed

Well  
Conditioned  
 $S_{xx}$  Matrix

$$\begin{bmatrix} S_{11} & S_{12} \\ S_{21} & S_{22} \end{bmatrix} \begin{bmatrix} H_1 \\ H_2 \end{bmatrix} = \begin{bmatrix} S_{1y} \\ S_{2y} \end{bmatrix}$$

Transfer Paths

$S_{xy}$  Matrix

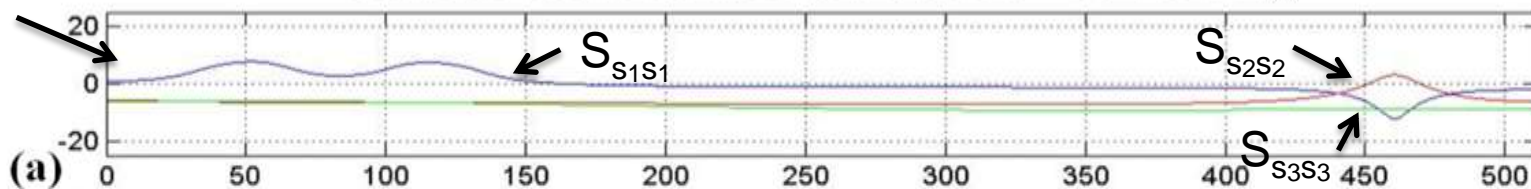
- Solution of this matrix by elementary row operations (Gaussian elimination) was conducted;

$$\begin{bmatrix} S_{11} & S_{12} \\ 0 & S_{22} - \frac{S_{12}S_{21}}{S_{11}} \end{bmatrix} \begin{bmatrix} H_1 \\ H_2 \end{bmatrix} = \begin{bmatrix} S_{1y} \\ S_{2y} - \frac{S_{1y}S_{21}}{S_{11}} \end{bmatrix}$$

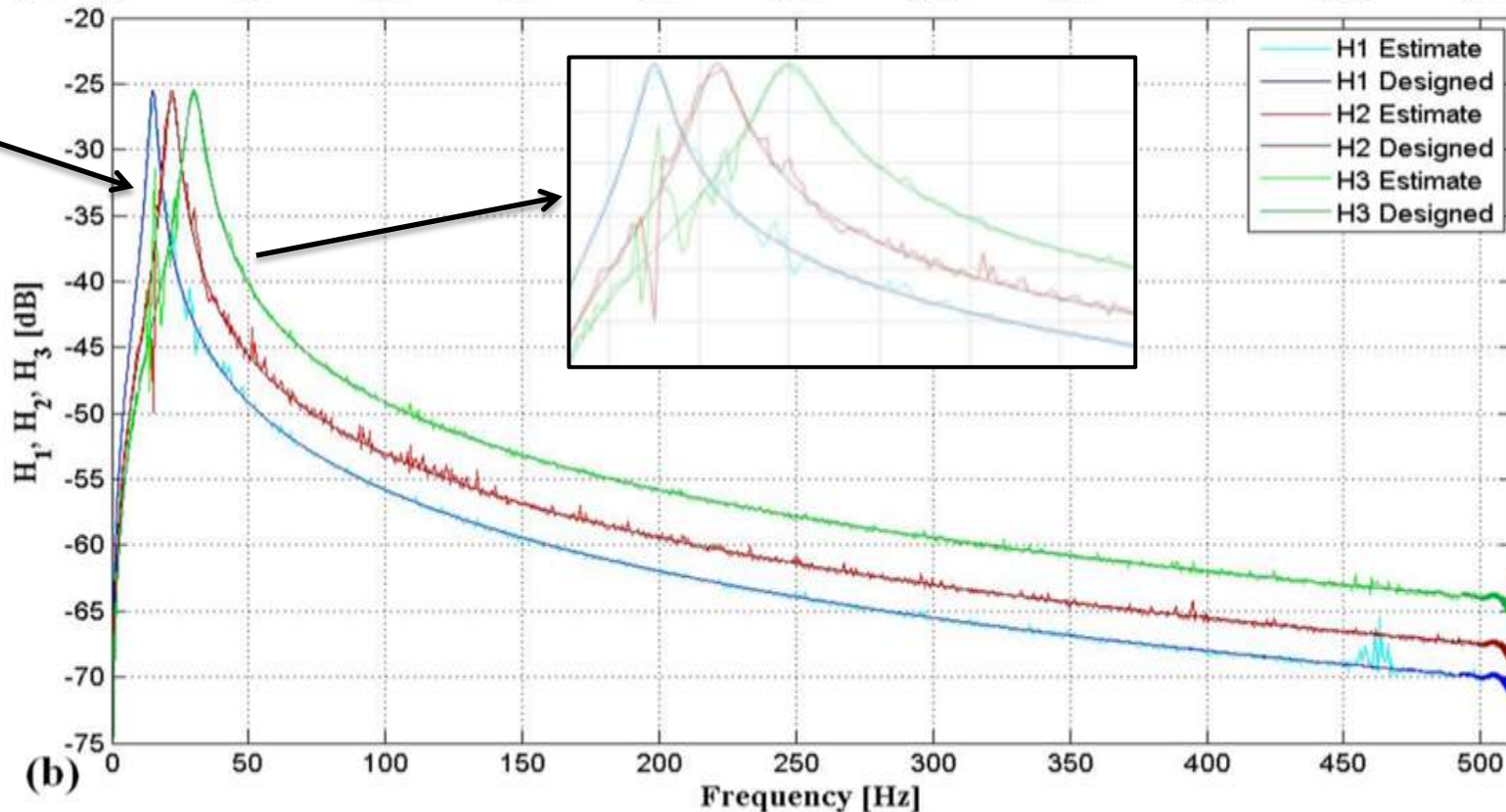
# Transfer Path Estimation

Independent source spectral characteristics

Transfer Path Estimate Simulation – Full Mixing

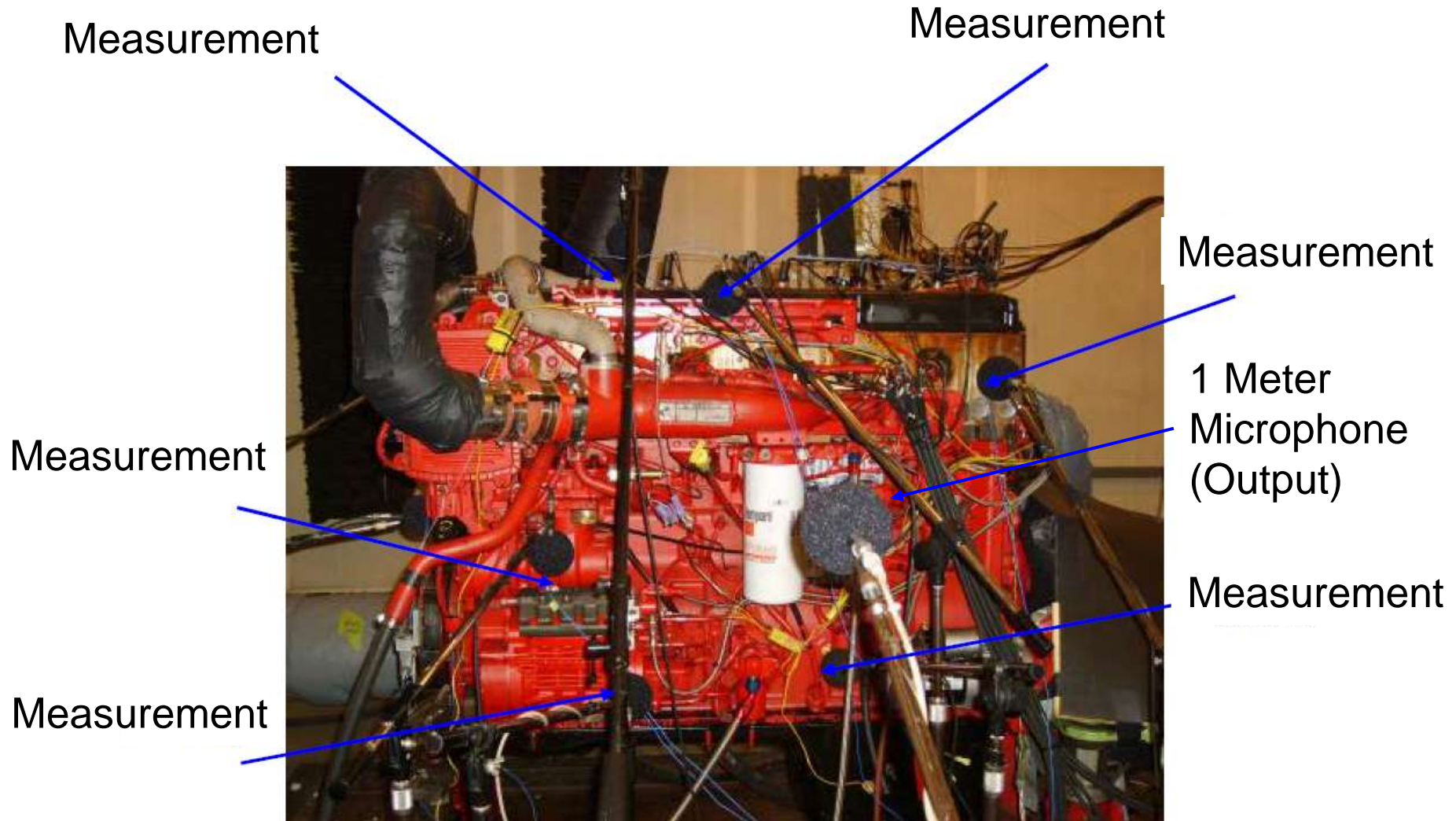


Designed and estimated transfer paths





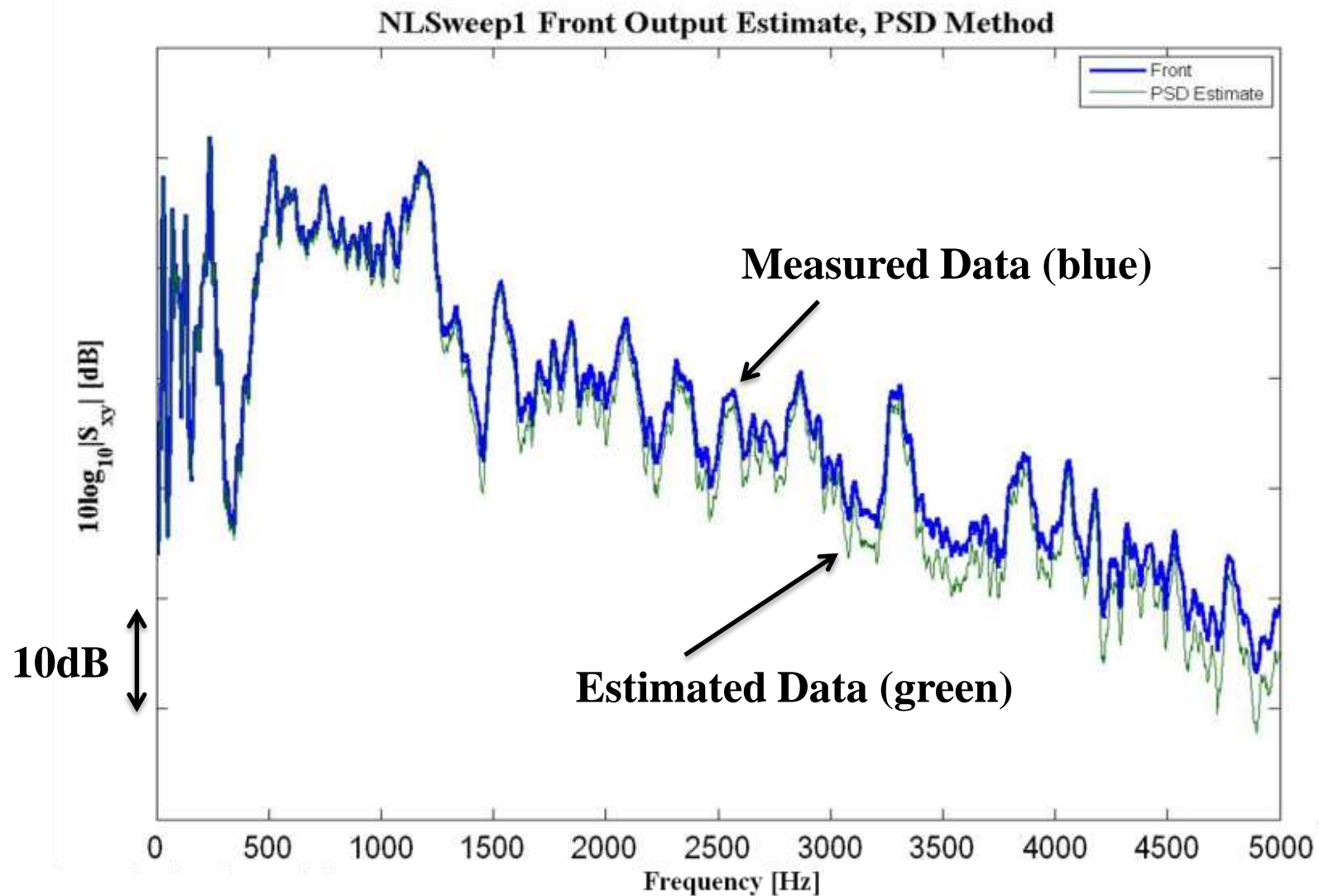
# Testing of Engine



- Each near-field measurement is an accelerometer/near-field microphone pair
- There are also 6 cylinder pressure transducers and 4 far-field microphones



# Far Field Estimate Power Spectra



# Singular Value Decomposition

Separate noise from uncorrelated sources from the data measured by the inputs

$$\mathbf{S}_{xx} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}^H = [\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_N] \text{diag}[\lambda_1, \lambda_2, \dots, \lambda_N] [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_N]^H$$

$$U = V$$

$$\lambda_1 > \lambda_2 > \dots > \lambda_N$$

$[\mathbf{S}_{xx}]$  = Cross spectral matrix

$N$  = Number of input

measurements to system

$\lambda_i$  = Singular value

$\mathbf{u}_i$  = Left singular vector

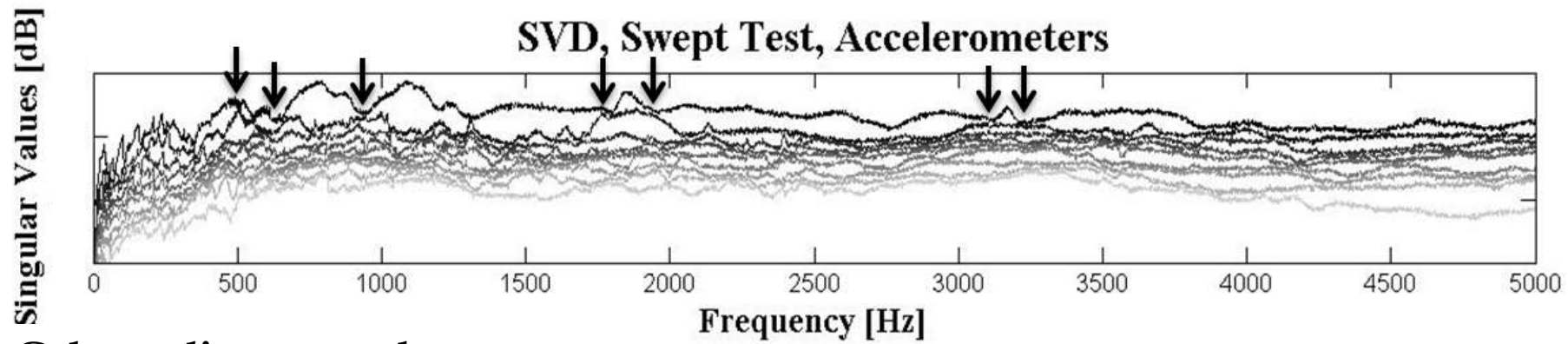
$\mathbf{v}_i$  = Right singular vector

$\mathbf{u} = \mathbf{v}$  in this case

Conducting a decomposition on the accelerometer or near field microphone spectral density matrices will help us identify the number of uncorrelated sources being measured, and potentially their relative strength.

Golub G H and Loan C F V 1996 *Matrix Computations*, 3<sup>rd</sup> Edition, (Baltimore: Johns Hopkins University Press).

# SVD Example and Decomposition



## Color coding example

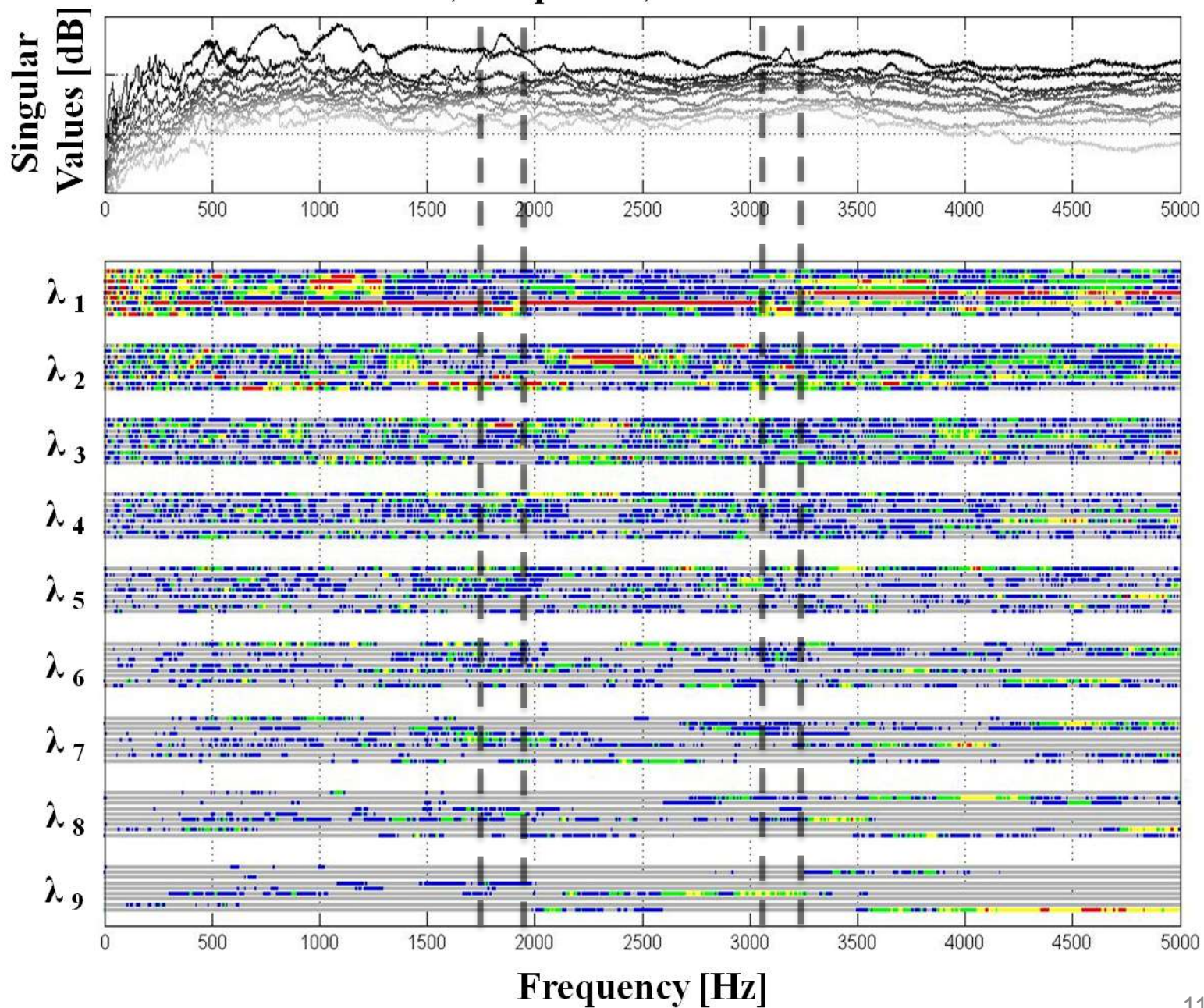
	1 <sup>st</sup> Singular Value Contribution	2 <sup>nd</sup> Singular Value Contribution	$n^{\text{th}}$ Singular Value Contribution
%			
>75			
50-75			
25-50			
5-25			
0-5			

$$S_{x_1x_1} = \underbrace{\mathbf{u}_1 \lambda_1 \mathbf{v}_1^H}_{\text{8\%}} + \underbrace{\mathbf{u}_2 \lambda_2 \mathbf{v}_2^H}_{\text{78\%}} + \dots + \underbrace{\mathbf{u}_n \lambda_n \mathbf{v}_n^H}_{\text{...}}$$

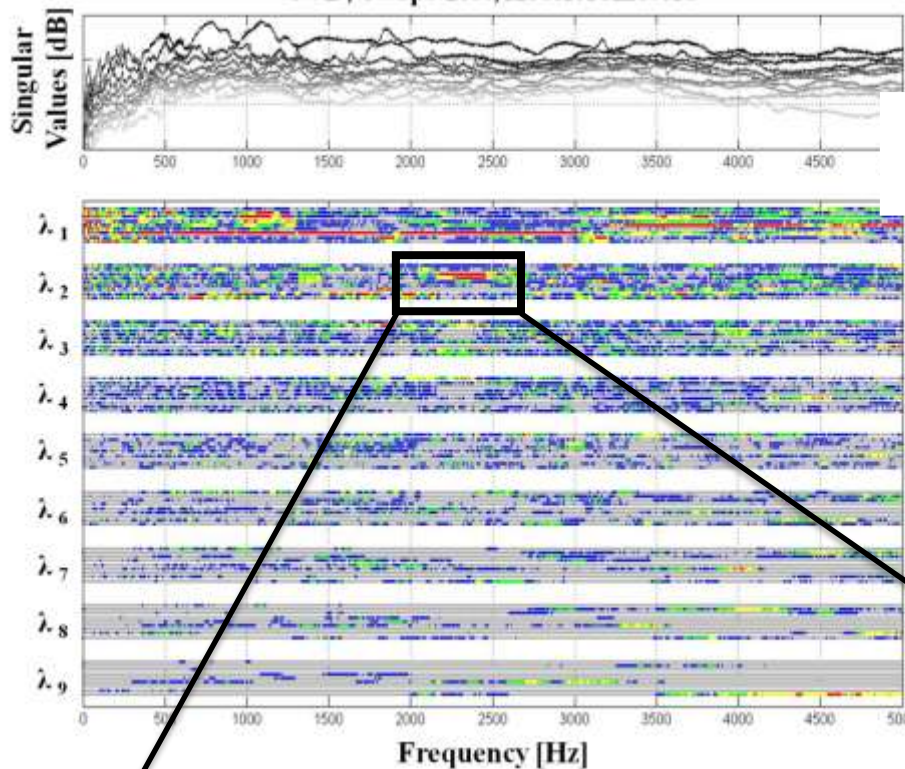
- Using a color coding scheme depending on the percentage contribution, these can be visualized graphically.
- A more detailed description of this method can be found in Hayward, Bolton & Davies (2012).

## SVD, Swept Test, Accelerometers





# SVD, Swept Test, Accelerometers



%

>75

50-75

25-50

5-25

0-5

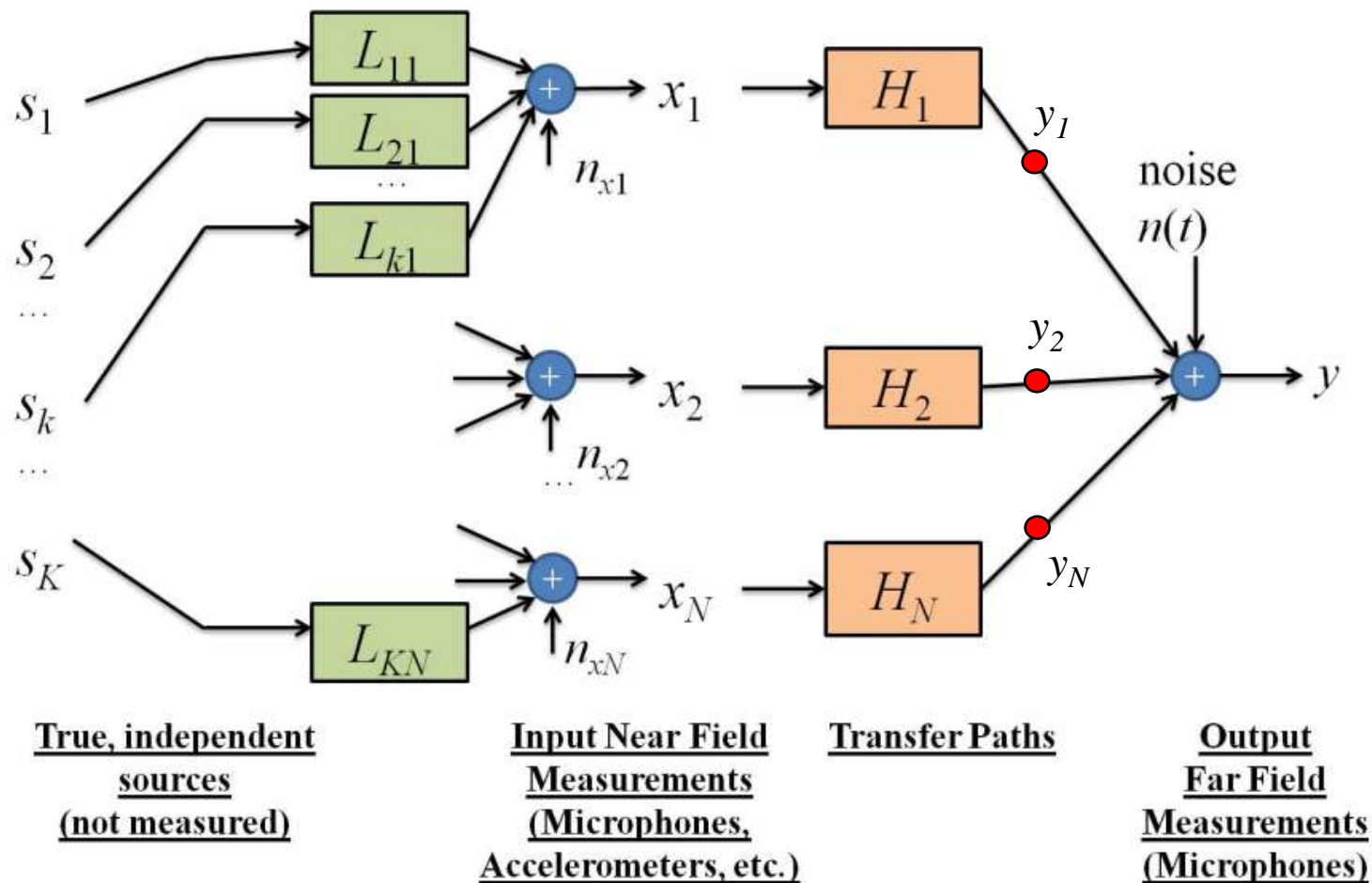
Note: This analysis uses input measurement data – contribution of singular values at output is not used

2<sup>nd</sup> Singular Value Contribution to:

- Measurement 1
- Measurement 2
- Measurement 3
- Measurement 4
- Measurement 5
- Measurement 6
- Measurement 7
- Measurement 8
- Measurement 9

Frequency [Hz]

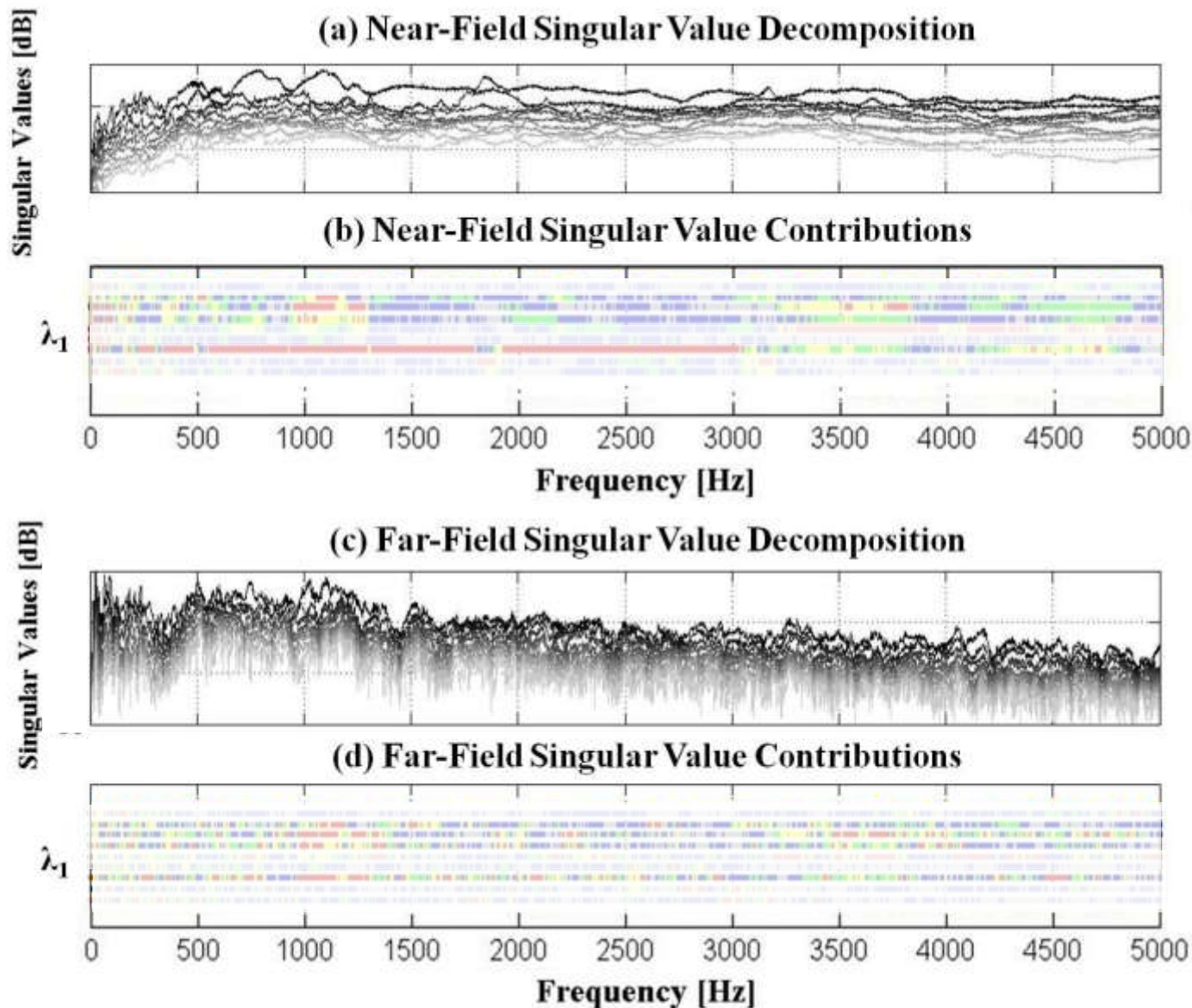
# Far Field Source Contributions



With calculated transfer paths, far-field time histories can be calculated, and source contributions to the far field can be determined

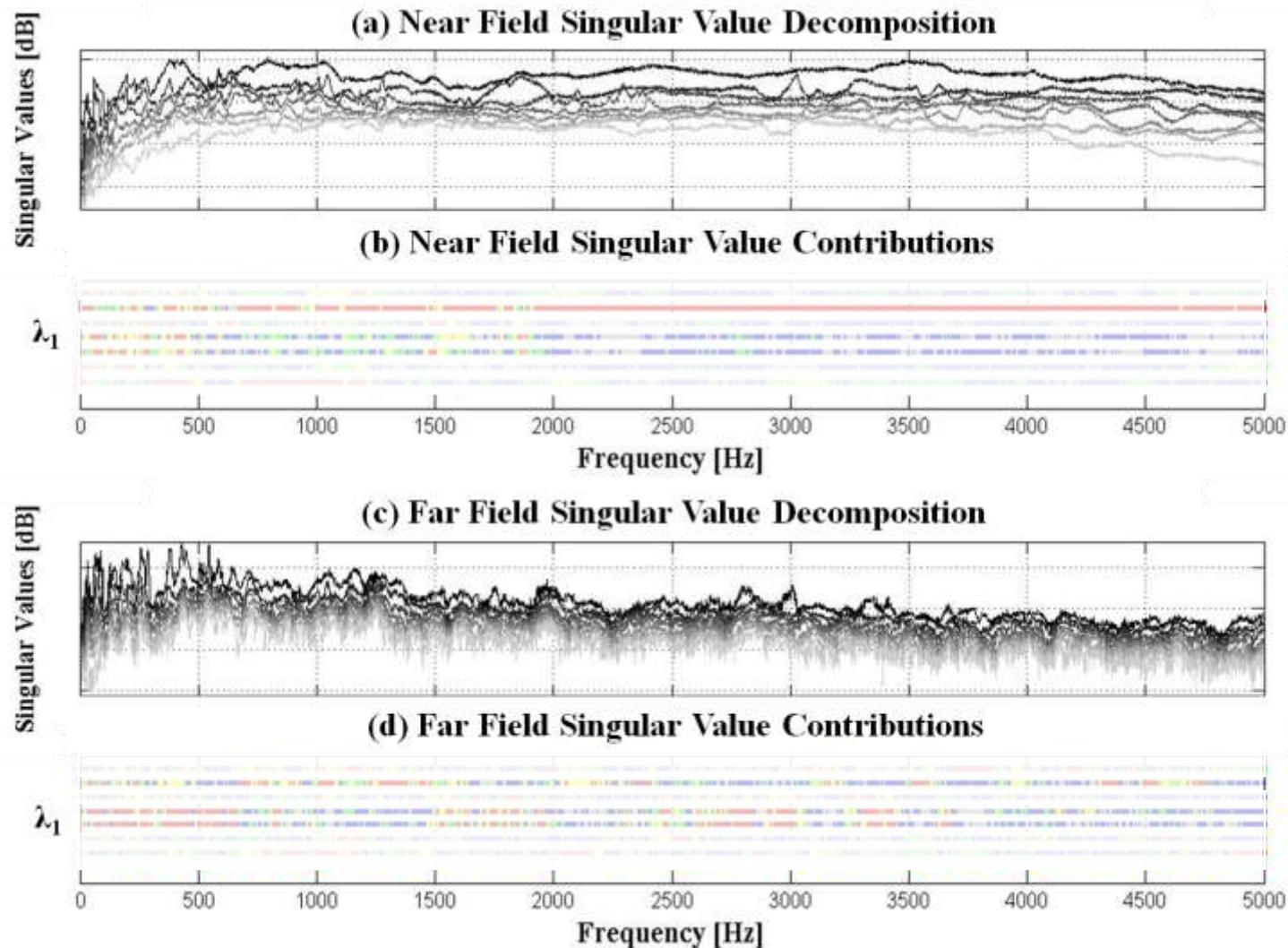


# Near-Field and Far-Field Contribution Comparison Example 1



# Near-Field and Far-Field Contribution Comparison Example 2

The analysis below is was conducted on data from a different engine, in a full loaded sweep test, with a different set of transducers



# Conclusions

- Transfer paths from input to output measurements can be accurately estimated through solution of a cross-spectral matrix problem.
- Contributions of each independent virtual source to real, physical near-field locations can be determined through singular value contribution plots.
- Utilization of input measurements and estimated transfer paths yield both accurate far-field estimate time histories, and SVD contribution plots demonstrating virtual source contributions in the far-field.
- Analysis of the singular values and their contributions to near- and far-field measurement power spectra allows inferences to be drawn regarding the characteristics of dominant noise sources within the engine.

# References

- Otte D, Sas P and Van de Ponsele 1988 *Noise Source Identification by use of Principal Component Analysis*, Proceedings of Inter-Noise 88 (France: Anvignon)
- Kompella MS, Ufford DA, Davies P and Bernhard RJ 1996 *A technique to determine the number of incoherent sources contributing to the response of a system*, Mechanical Systems and Signal Processing, vol. 8 no. 4 pp. 363-380
- Leclère Q, Pèzerat C, Laulagnet B and Polac L 2005 *Application of Multi-Channel Spectral Analysis to Identify the Source of a Noise Amplitude Modulation in a Diesel Engine Operating at Idle*, Applied Acoustics, vol. 66 no. 7 pp. 779-798
- Hayward, M.D., Bolton, J.S., and Davies, P. 2012 *Connecting the singular values of an input cross-spectral density matrix to noise sources in a diesel engine*, INTER-NOISE and NOISE-CON Congress and Conference Proceedings, Vol. 2012, no. 2, pp. 9583-9593, Institute of Noise Control Engineering

## Singular Value Decomposition

- Golub G H and Loan C F V 1996 *Matrix Computations*, 3<sup>rd</sup> Edition, (Baltimore: Johns Hopkins University Press).